



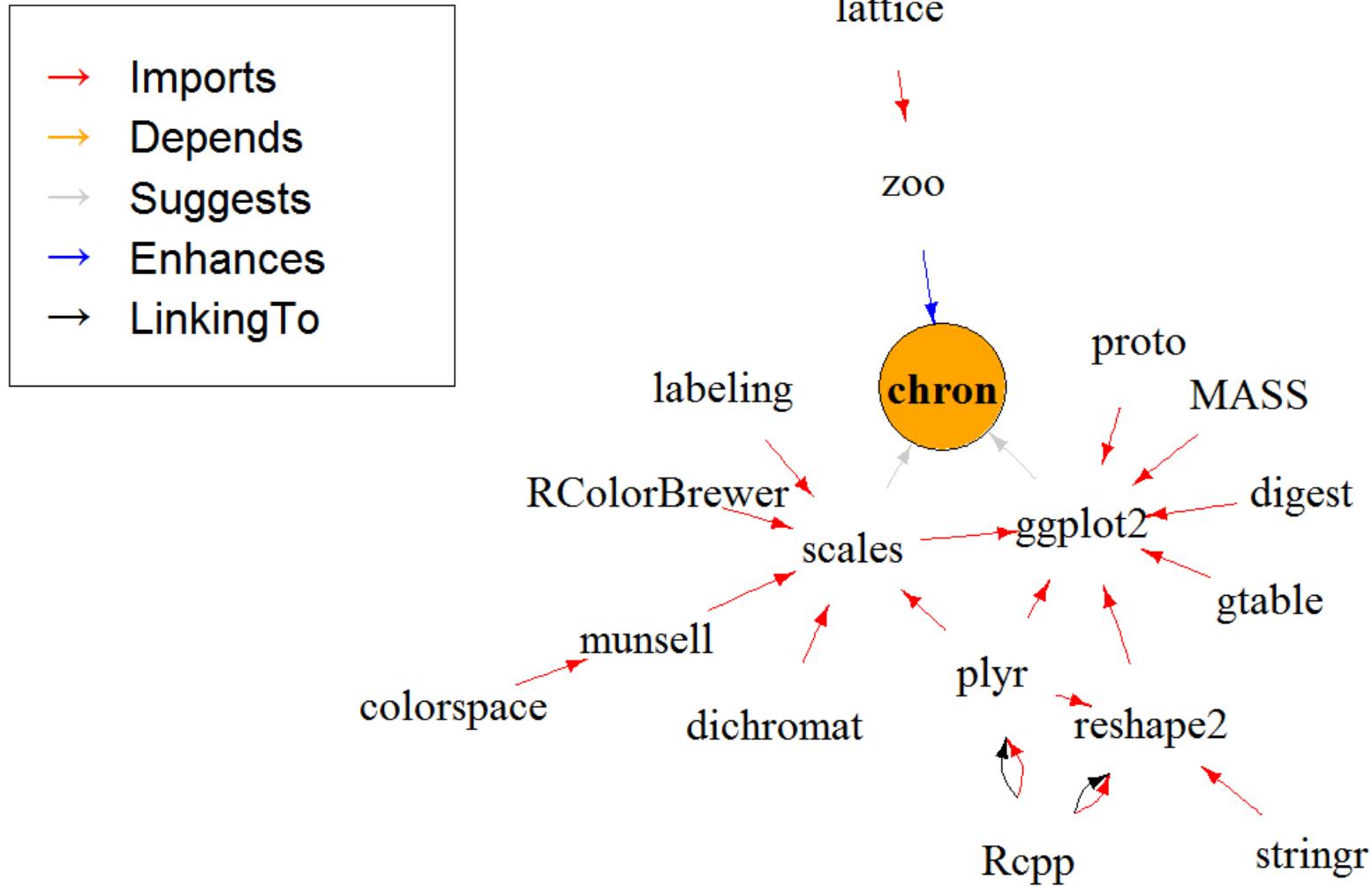
The network structure of CRAN

Andrie de Vries

adevries@microsoft.com
@RevoAndrie

UseR!2015, Aalborg

A network of package dependencies



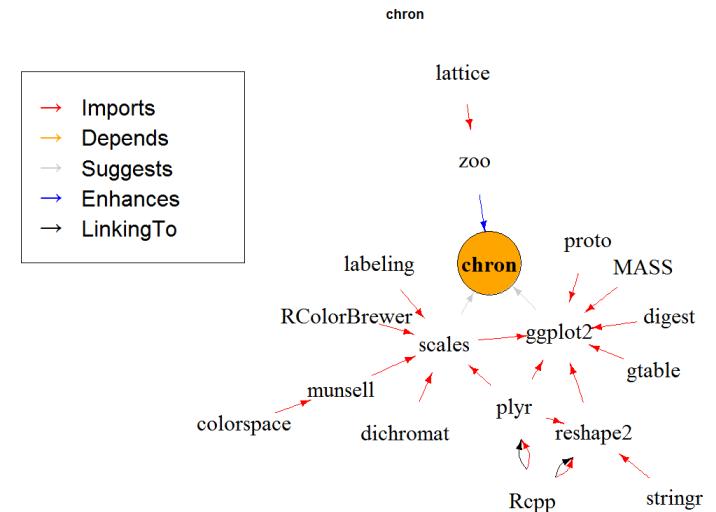
Using miniCRAN to create the network graph

igraph

→

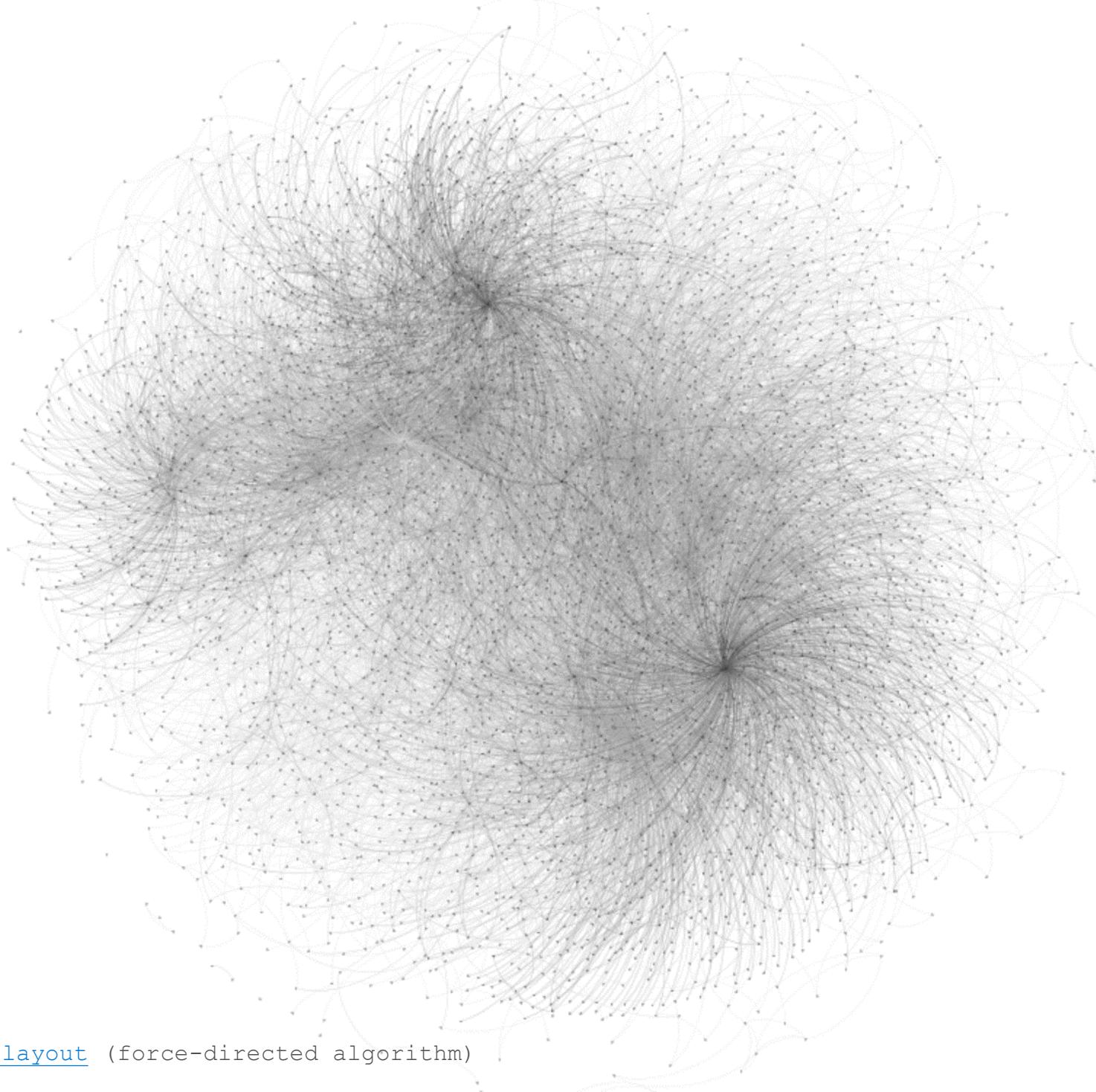
miniCRAN

```
pdb <- available.packages()
tags <- "chron"
library("miniCRAN")
dg <- makeDepGraph(tags, availPkgs = pdb)
plot(dg)
```



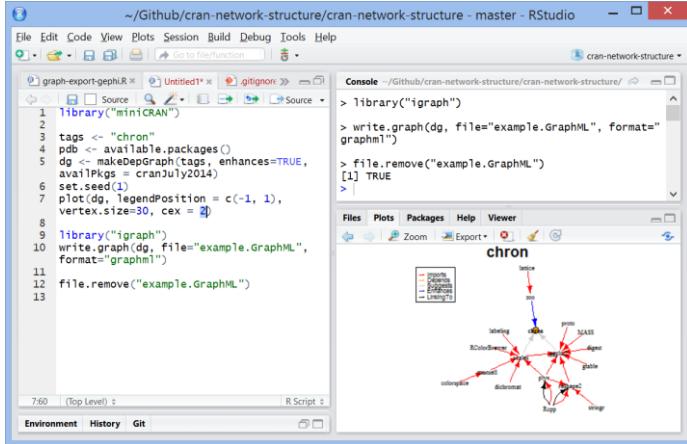
How to visualize a network of 6,791 CRAN packages?

CRAN

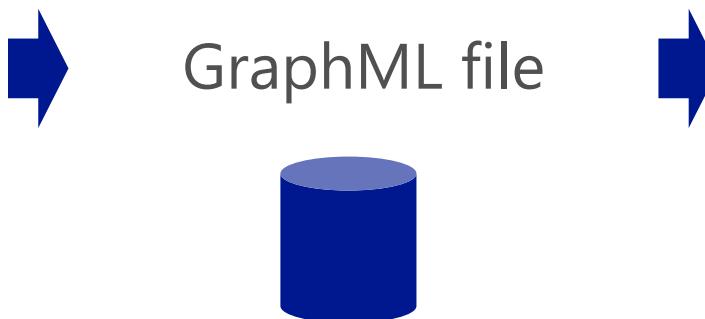


Note: [Fruchterman Reingold layout](#) (force-directed algorithm)

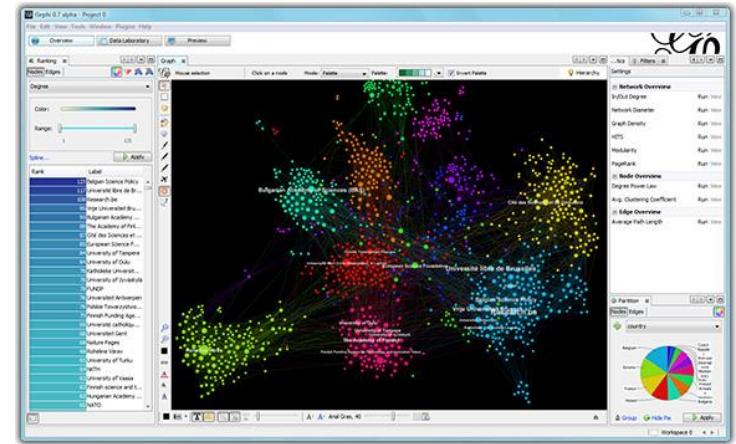
How to make this plot



```
~/GitHub/cran-network-structure/cran-network-structure - master - RStudio
File Edit Code View Plots Session Build Debug Tools Help
cran-network-structure
graph-export-gephi.R  Untitled1* gitignore Source
1 library("igraph")
2
3 tags <- "chron"
4 pdbs <- available.packages()
5 dg <- makeDepGraph(tags, enhances=TRUE,
6 availPkgs = cranJuly2014)
7 set.seed(1)
8 plot(dg, legendPosition = c(-1, 1),
9 vertex.size=30, cex = 3)
10 write.graph(dg, file="example.GraphML",
11 format="graphml")
12 file.remove("example.GraphML")
13
```



GraphML file



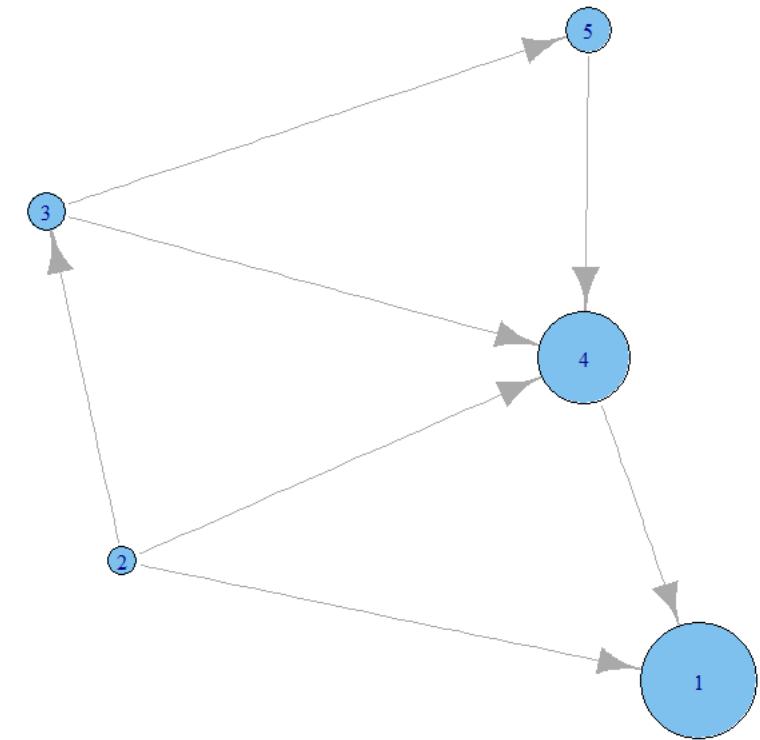
Gephi
(gephi.github.io)

```
library("igraph")
write.graph(dg,
  file = "example.GraphML",
  format = "graphml")
```

Which nodes (packages) are more “important”?

Reverse dependencies influence importance

- Imagine that each package dependency is a vote of confidence, cast by the community of package authors
- How can we compute the importance of a package to downstream packages?

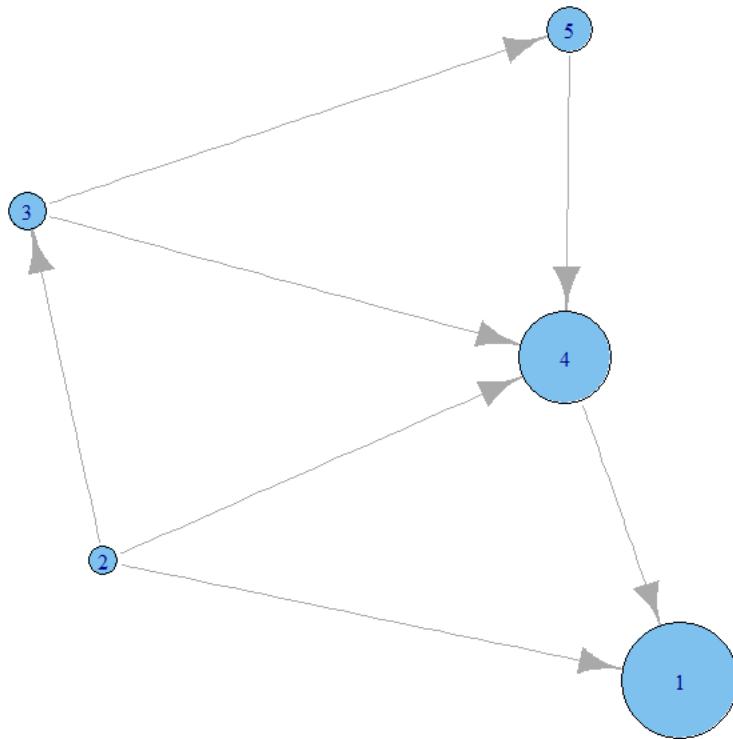


Page rank

- Page rank is famous as one of the criteria in google search results
- You can use page rank on any graph to compute “importance”

```
igraph::page.rank()
```

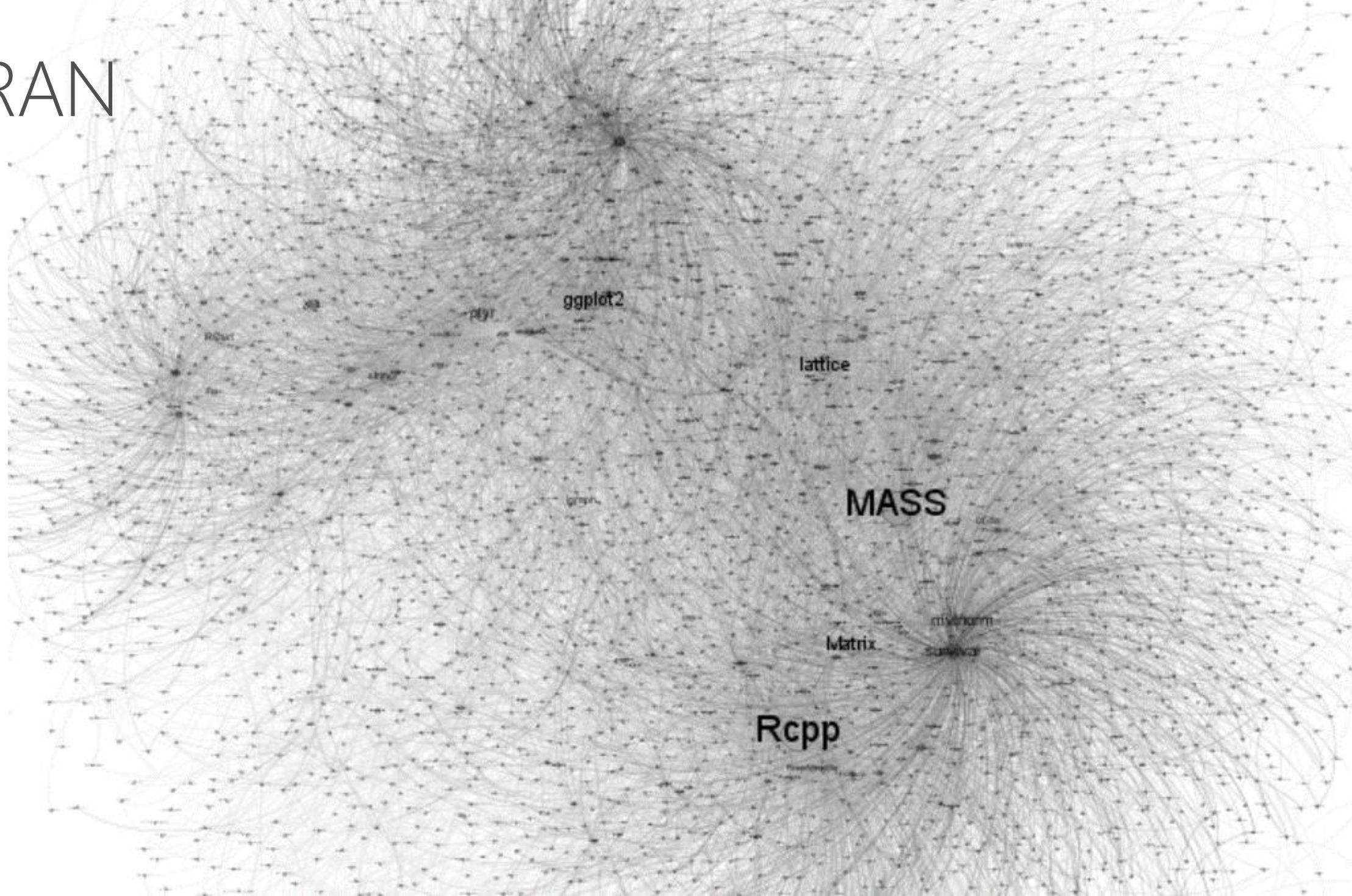
Simple graph with vertex size proportional to page.rank



Pagerank results

<u>Package</u>	<u>PageRank</u> <u>(Dec 2014)</u>	<u>PageRank</u> <u>(Jun 2015)</u>	<u>Description</u>
Rcpp	0.0166	0.0197	Interface to use C++ code in R
MASS	0.021	0.0196	Functions and datasets to support Venables and Ripley, 'Modern Applied Statistics with S' (4th edition, 2002).
Matrix	0.01	0.0095	Sparse matrix engine
ggplot2	0.0073	0.0086	Graphics engine
lattice	0.0096	0.0085	Base R package for lattice (trellis) graphics
mvtnorm	0.0088	0.0083	Multivariate normal distributions
survival	0.0083	0.0079	Time-to-event analysis
plyr	0.0067	0.0072	Group-by operations
igraph	0.0047	0.0049	Analyse graph structures
XML	0.0047	0.0047	Parse and manipulate documents in XML format

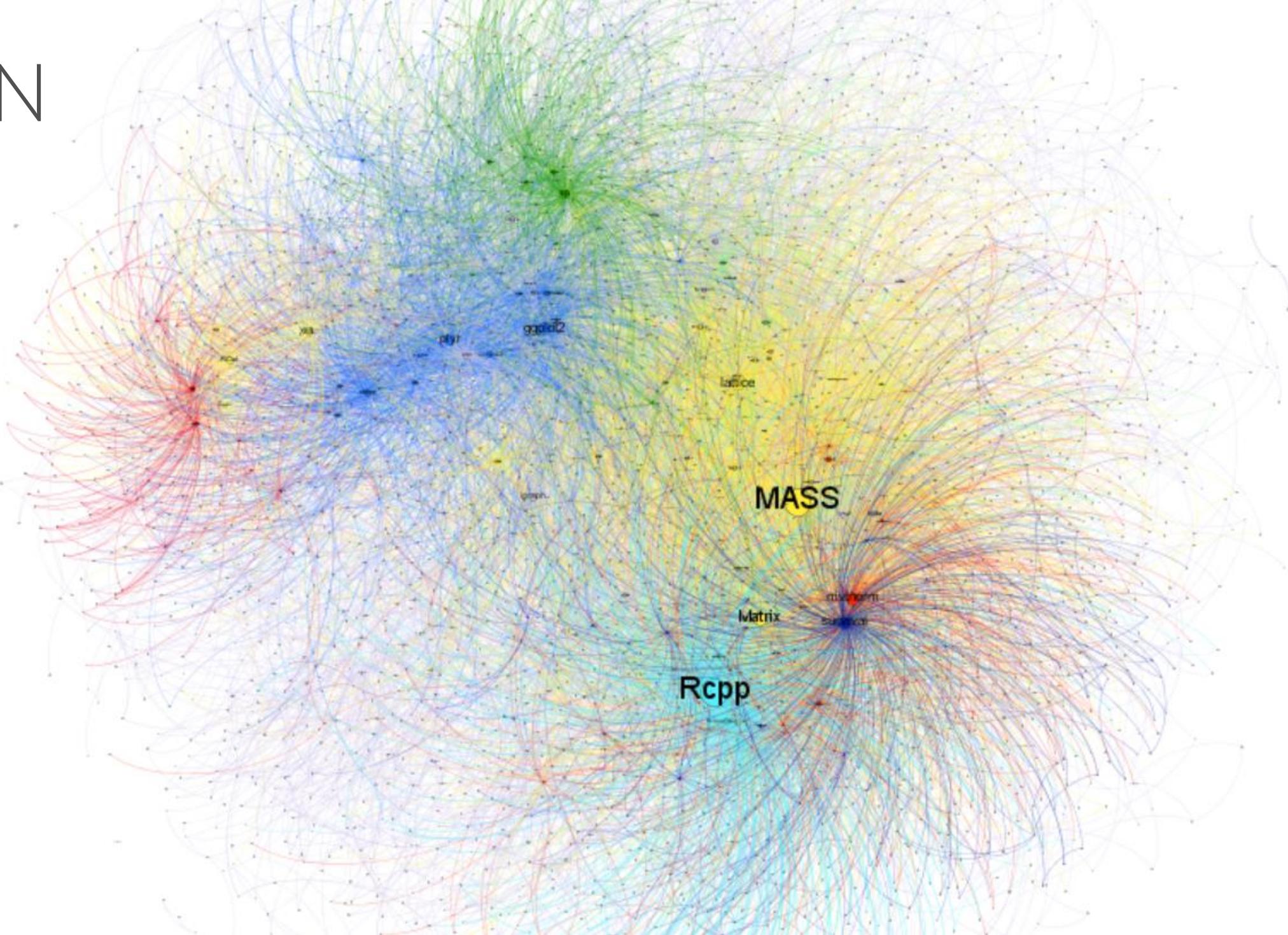
CRAN



Can we detect clusters of packages?

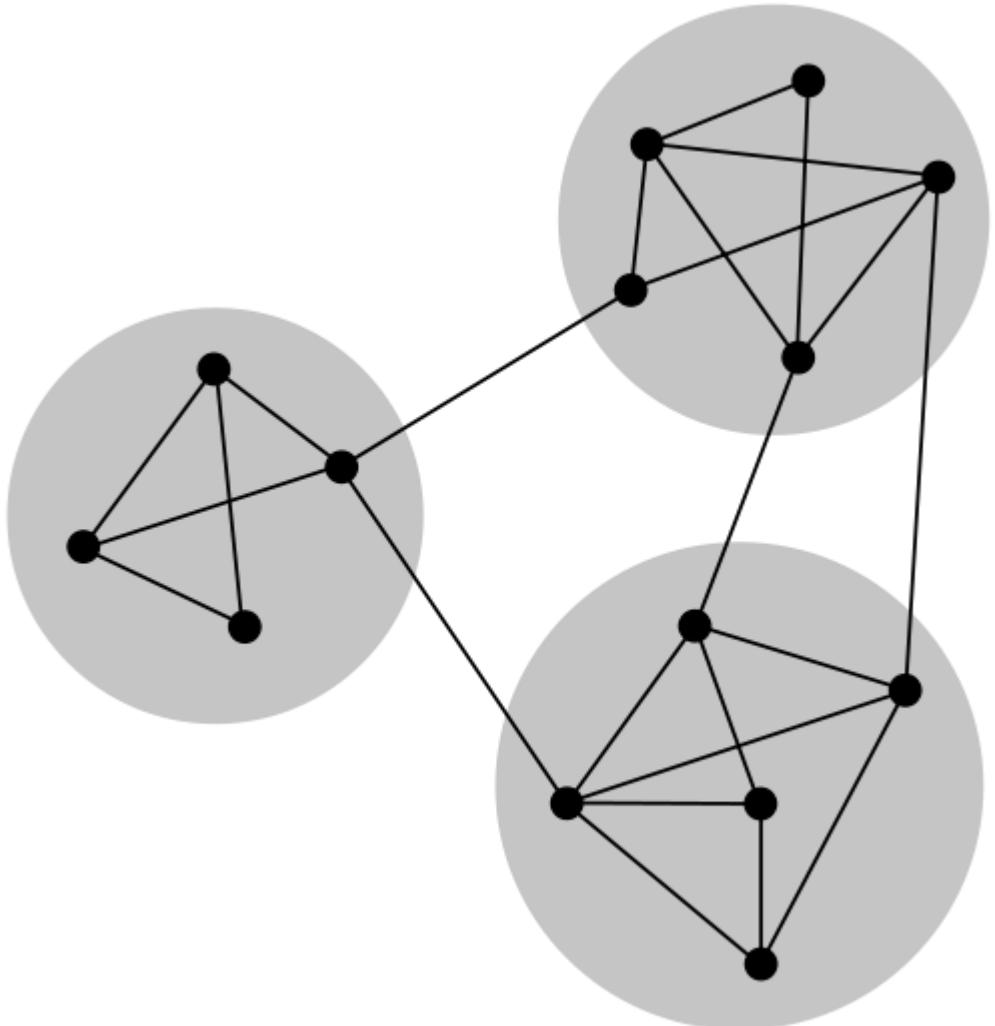
CRAN

CRAN



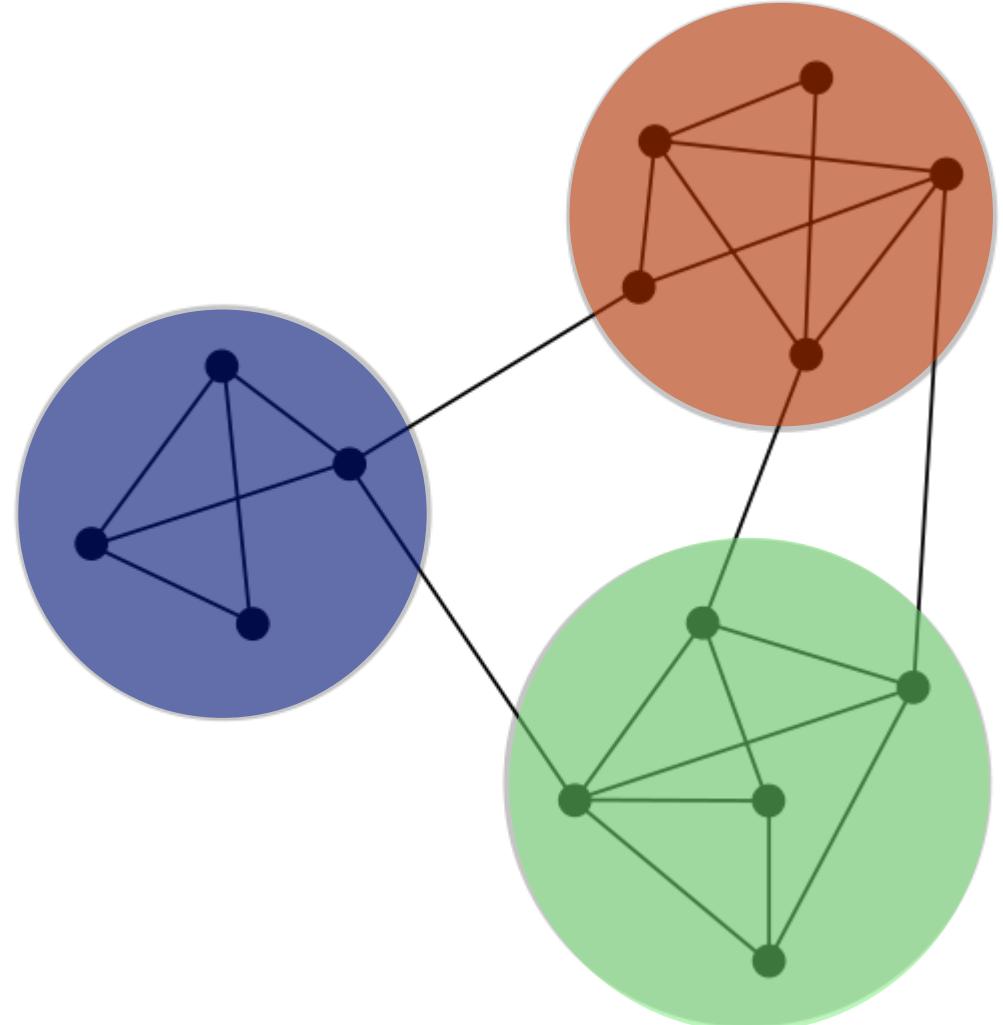
Community detection

- Communities are densely connected groups of nodes in a graph
- Several implemented algorithms in igraph:
 - Fast greedy
 - Walktrap
 - Spinglass
 - Leading eigenvector
 - Edge betweenness



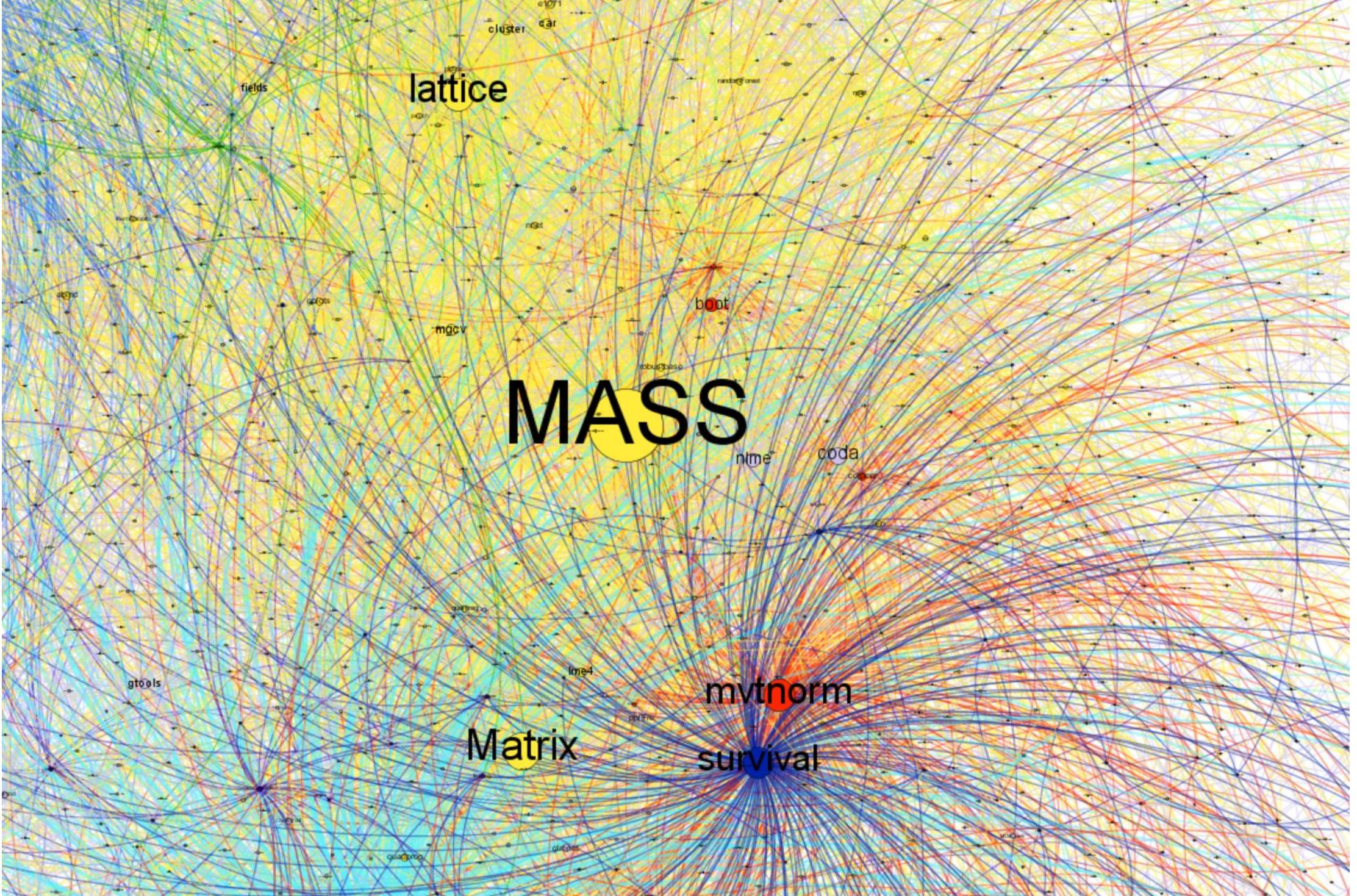
The walktrap algorithm

- `?walktrap.community`
 - This function tries to find densely connected subgraphs, also called communities in a graph via random walks. The idea is that short random walks tend to stay in the same community.
 - Efficient algorithm for large, densely connected graphs*



CRAN

Zooming in



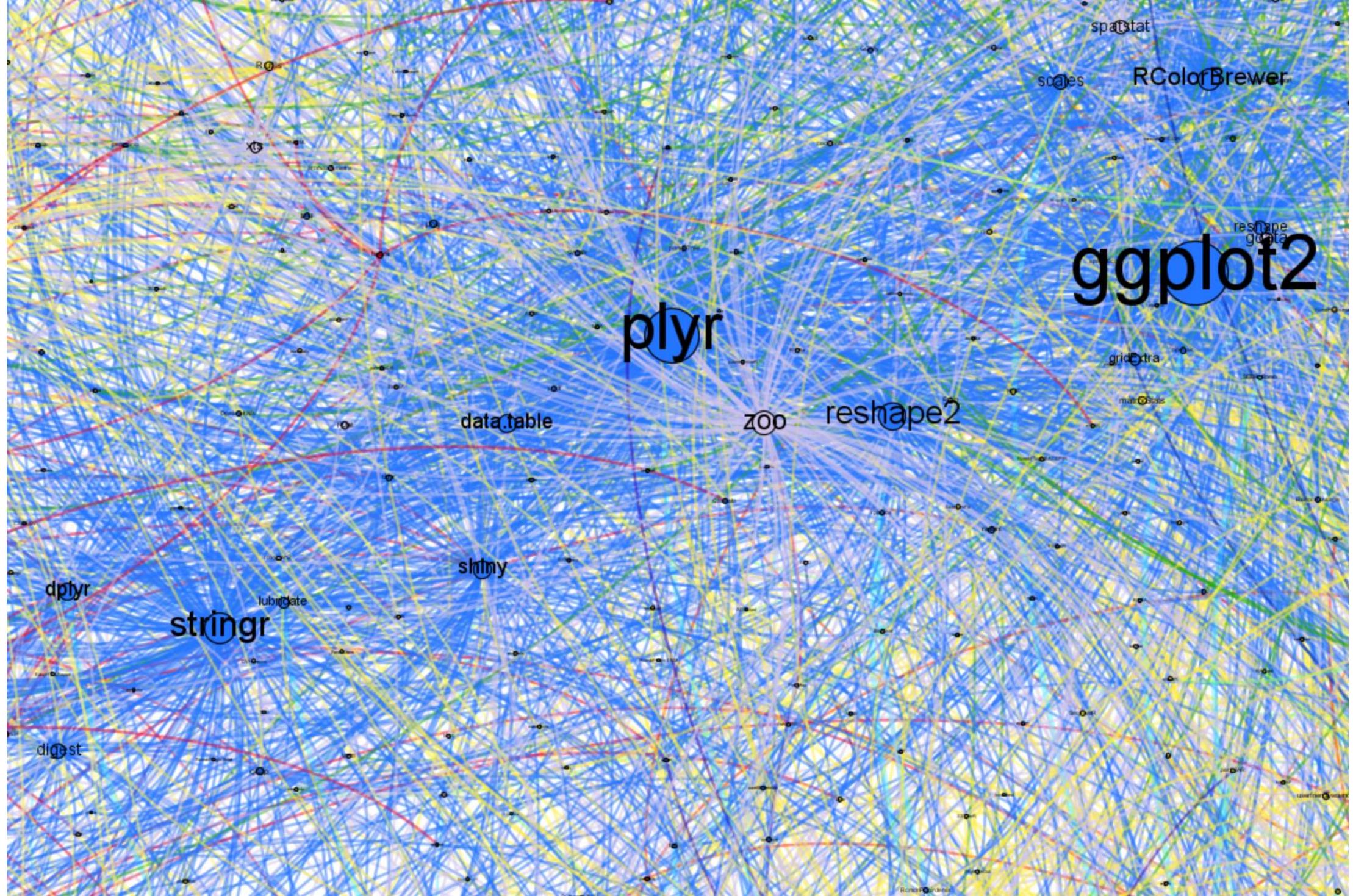
lattice

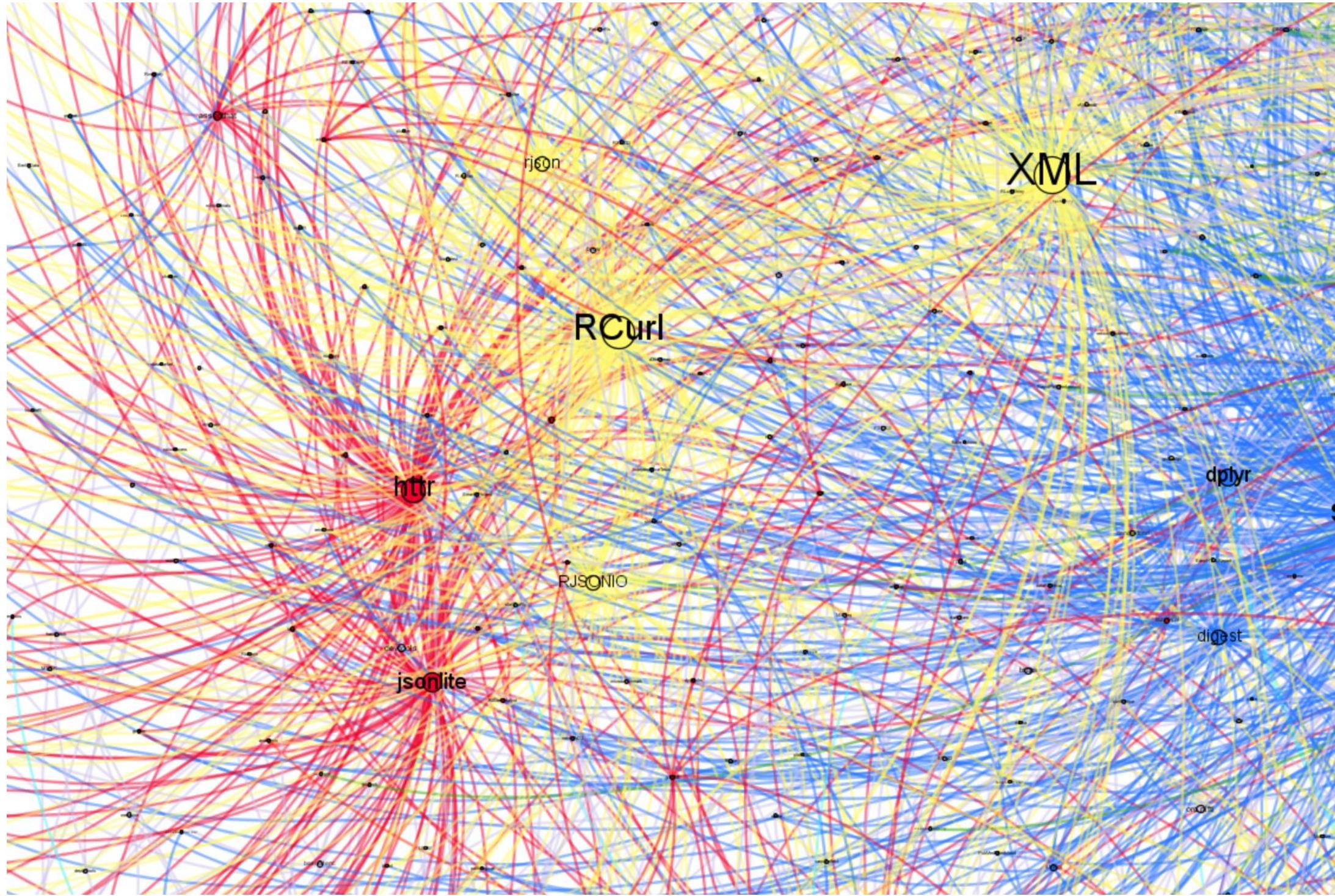
MASS

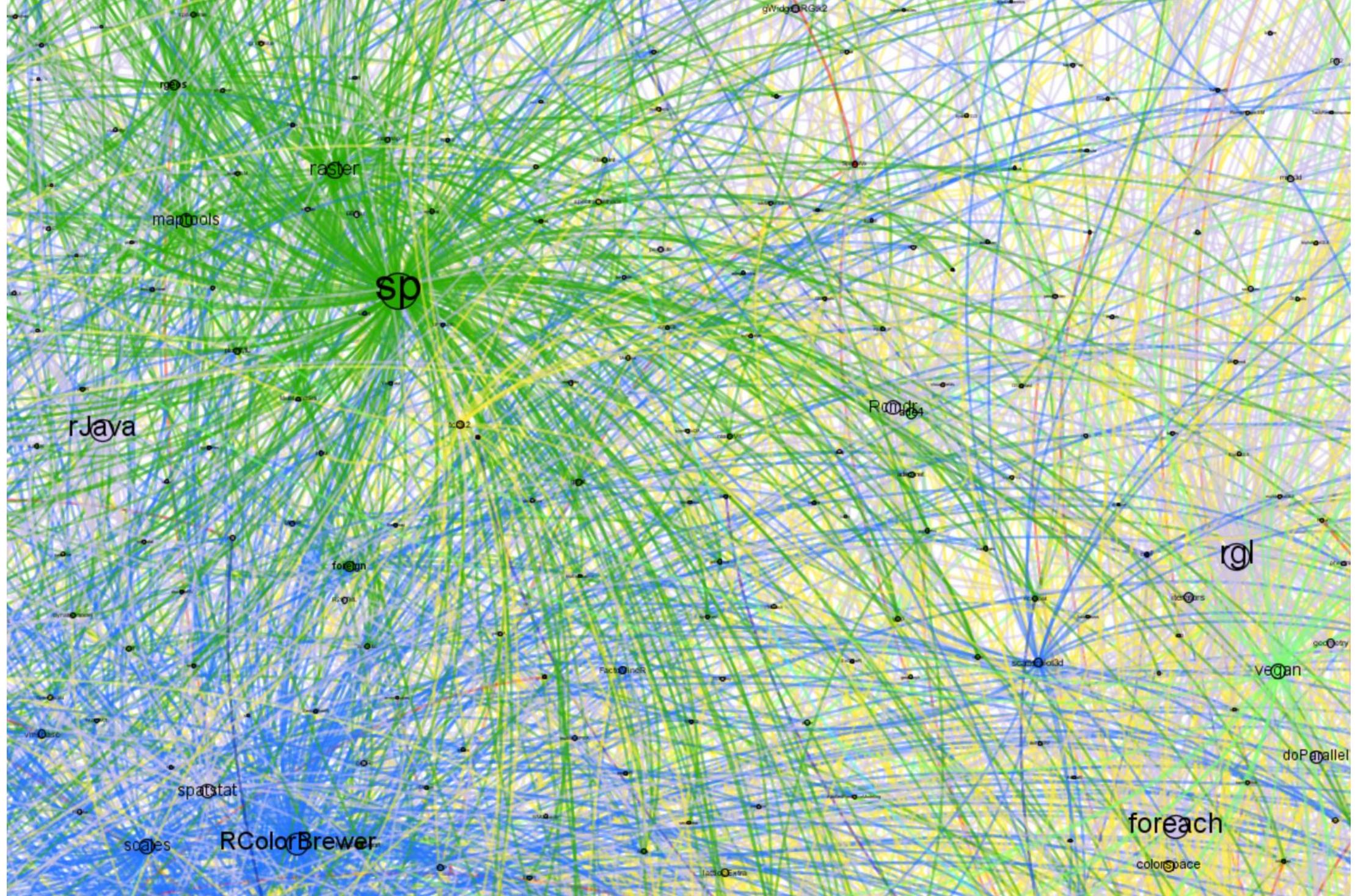
Rcpp

Matrix

mytnorm
survival

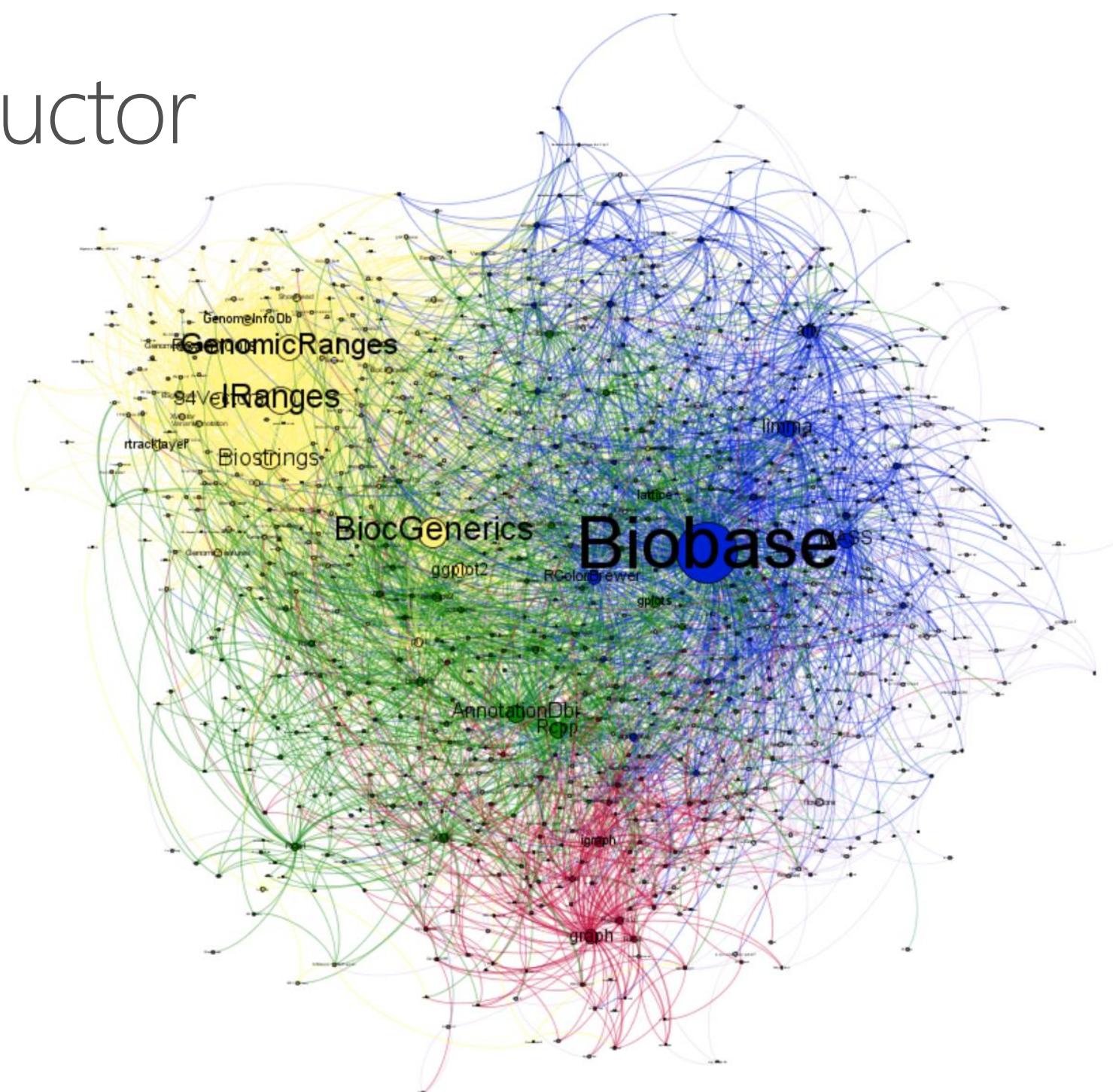






Is BioConductor different?

BioConductor



Further analysis required

Summary

Summary

```
library(miniCRAN)
```

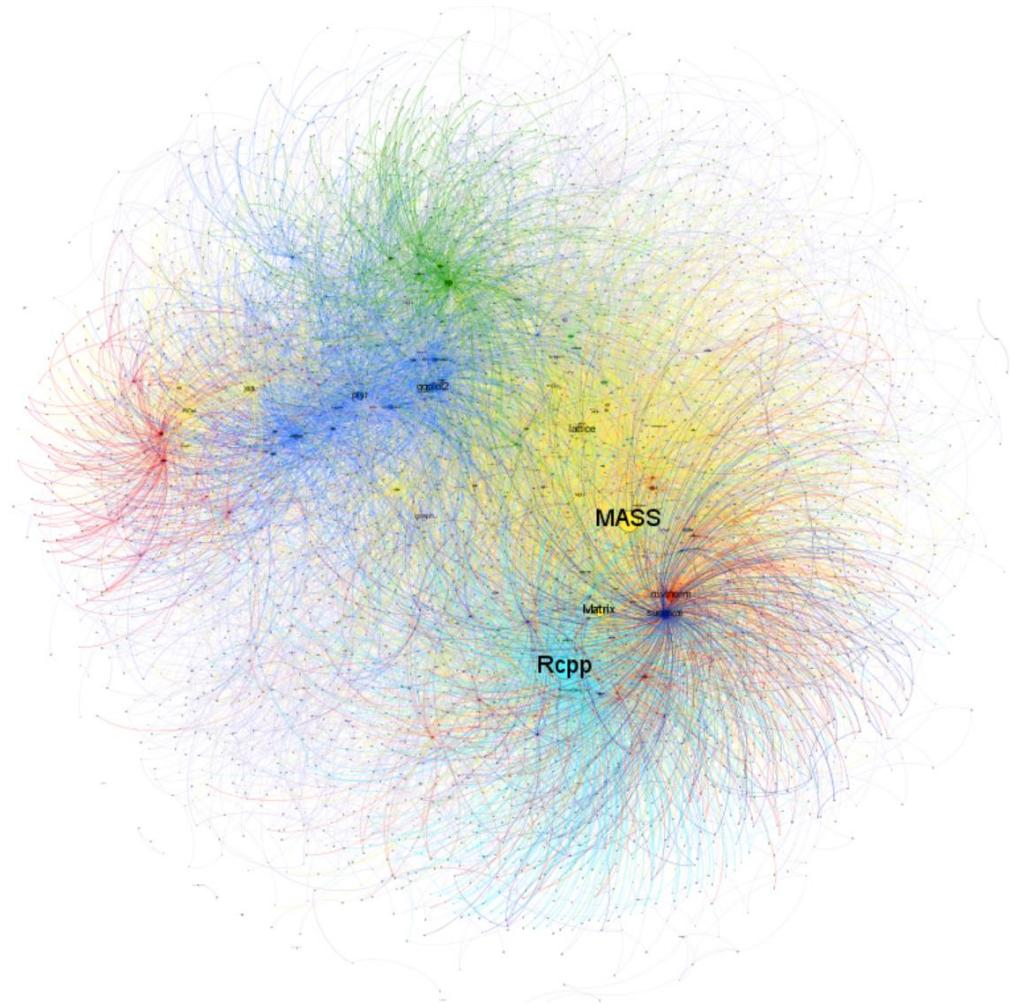
```
makeDepGraph()
```

```
library(igraph)
```

```
page.rank()
```

```
walktrap.community()
```

```
write.graph()
```



Scripts available at:

<https://github.com/andrie/cran-network-structure>

adevries@microsoft.com

@RevoAndrie

